

Quantum Algorithms for Composed Functions With Shared Inputs



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Introduction

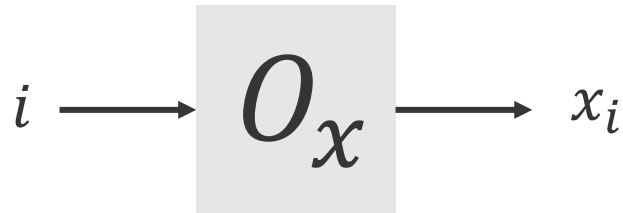
Query complexity

Let $f: \{0,1\}^n \rightarrow \{0,1\}$ be a function and $x \in \{0,1\}^n$ be an input to f .

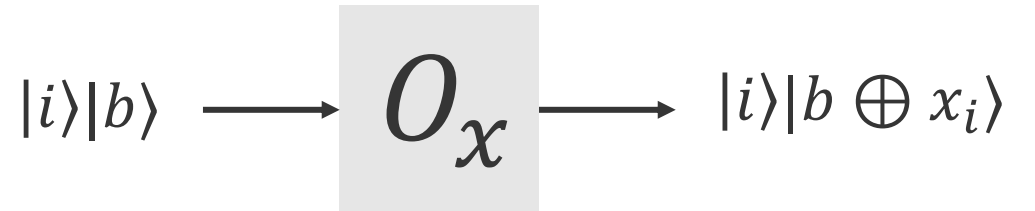
$$x = \begin{array}{|c|c|c|c|c|} \hline x_1 & x_2 & x_3 & \cdots & x_n \\ \hline \end{array}$$

Goal: Compute $f(x)$ by reading as few bits of x as possible.

Equivalently, compute $f(x)$ using a circuit/algorithm with the least number of uses of this oracle:



In the quantum setting, we have this oracle:



Why query complexity?

Quantum algorithmic motivation

- Query algorithms often can be made time-efficient, while the abstraction of query complexity often gets rid of unnecessary details.
- Most quantum algorithms are naturally phrased as query algorithms. E.g., Shor, Grover, Hidden Subgroup, Linear systems (HHL), etc.

Classical algorithmic motivation

- Functions with quantum query complexity $< d$ can be agnostically PAC learned in time $\sim n^d$.

Classical complexity motivation

- If all circuits in a class C have quantum query complexity $o(n)$, then no circuit from C can correctly compute IP2 on a $\frac{1}{2} + 1/\text{poly}(n)$ fraction of inputs.

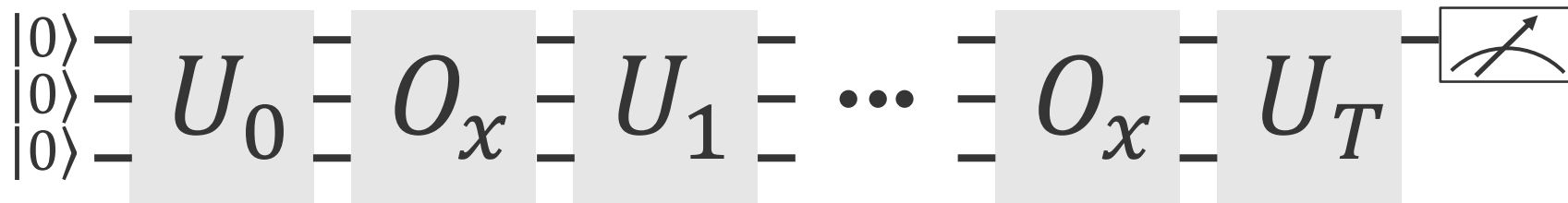
Other applications

- Oracle separations between classes, lower bounds on restricted models, upper and lower bounds in communication complexity, data structures, etc.

Quantum query complexity

Quantum query complexity: Minimum number of uses of O_x in a quantum algorithm that for every input x , outputs $f(x)$ with error $\leq 1/3$.

$Q(f)$



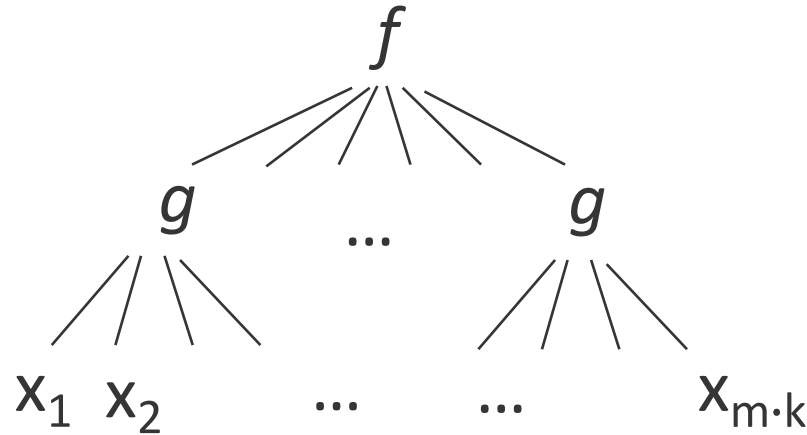
Example: Let $\text{OR}_n(x) = \bigvee_{i=1}^n x_i$ and $\text{AND}_n(x) = \bigwedge_{i=1}^n x_i$.

Then $Q(\text{OR}_n) = Q(\text{AND}_n) = \Theta(\sqrt{n})$ [Grover96, Bennett-Bernstein-Brassard-Vazirani97]

Classically, we need $\Theta(n)$ queries for both problems.

Query Complexity of Block-Composed Functions

- If $f: \{0,1\}^m \rightarrow \{0,1\}$ and $g: \{0,1\}^k \rightarrow \{0,1\}$, the block composition $f \circ g: \{0,1\}^{m \cdot k} \rightarrow \{0,1\}$ is defined via $f \circ g = f(g, \dots, g)$, with each copy of g on a disjoint set of variables.

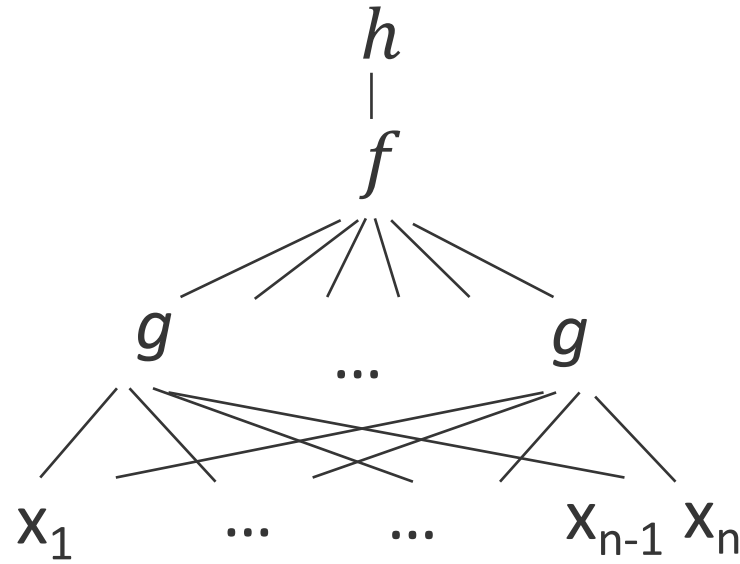


- Fact: $Q(f \circ g) = \Theta(Q(f) \cdot Q(g))$ [HLS 2007, Reichardt 2010].
- **Tomorrow:** Troy Lee's talk will discuss whether the analogous statement holds for randomized query complexity.

Main Composition Theorem

Query Complexity of Shared-Input Compositions

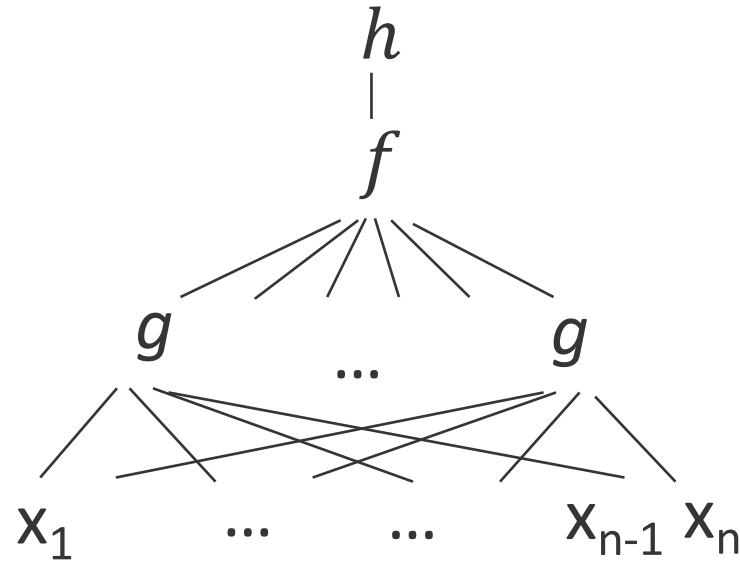
- We are interested in shared-input compositions.



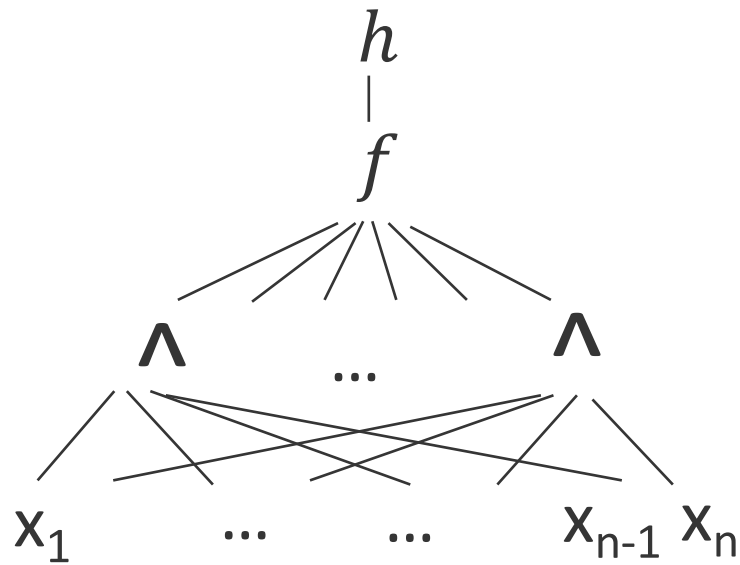
- A trivial upper bound for $Q(h)$ is: $Q(h) \leq O(Q(f) \cdot Q(g))$.
- Is it always possible to leverage the sharing of inputs to improve this upper bound?

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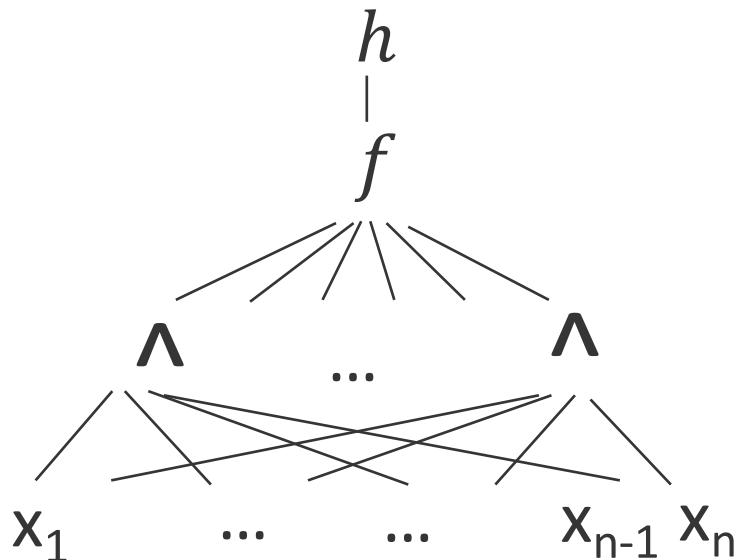


- A trivial upper bound for $Q(h)$ is: $Q(h) \leq O(Q(f) \cdot Q(g))$.
- Is it always possible to leverage the sharing of inputs to improve this upper bound?
- We show the answer is **YES** whenever $g=\text{AND}$, and $Q(f) \ll n$.



Main Theorem: $Q(h) \leq O(\sqrt{Q(f)} \cdot n \cdot \text{polylog}(n))$.

- Without leveraging input sharing, the best upper bound for $Q(h)$ is:
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- Without leveraging input sharing, the best upper bound for $Q(h)$ is:
 $Q(h) \leq O(Q(f) \cdot \sqrt{n}).$
- Also show an analog for **approximate degree**: $\text{adeg}(h) \leq \tilde{O}(\sqrt{\text{adeg}(f)} \cdot n).$
- Both results are tight for some functions, e.g., for $h = \text{PARITY}_t \circ \text{AND}_{\frac{n}{t}}.$

Idea of the Quantum Algorithm for h

- Query input bits with the goal of “killing” high fan-in AND gates.
 - i.e., if we query bit x_i and learn $x_i = 0$, then all AND gates involving x_i have their value fixed to 0, so we can effectively delete them.

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 - i.e., if we query bit x_i and learn $x_i = 0$, then all AND gates involving x_i have their value fixed to 0, so we can effectively delete them.
- Let h' denote h with all queried bits x_i restricted to their queried values.
- If all surviving AND gates have fan-in at most $n/Q(f)$, then the trivial upper bound yields:

$$Q(h') \leq Q(f) \cdot \sqrt{\frac{n}{Q(f)}} = \sqrt{n \cdot Q(f)}.$$

Idea of the Quantum Algorithm for h

- Remaining challenge: Figure out how to reduce the fan-in of all surviving AND gates to $n/Q(f)$, using only $\sqrt{n \cdot Q(f)}$ queries.
 - To accomplish this, we iteratively Grover search for an x_i such that:
 - $x_i = 0$ and
 - x_i is connected to “many” surviving high-fan-in AND gates.

Implications of the Composition Theorem

Quantum Query Algorithms for Linear-Size AC^0

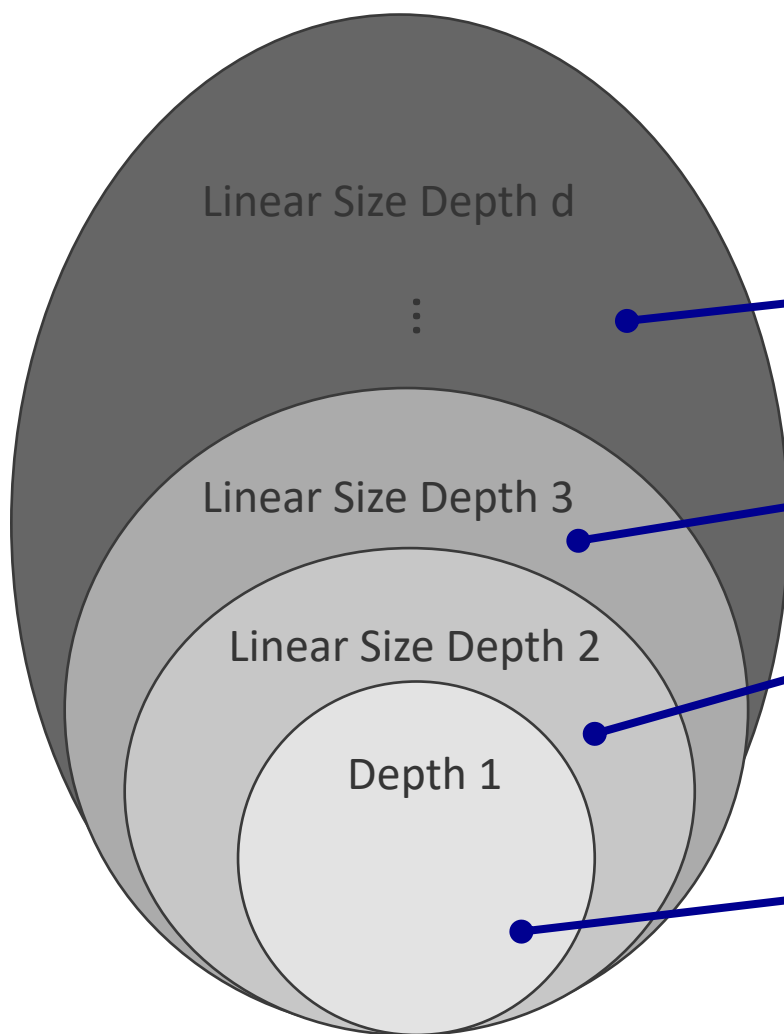
- Recursively applying our Main Theorem shows that any linear size, depth- d (i.e., LC_d^0) circuit f satisfies $Q(f) = \tilde{O}(n^{1-2^{-d}})$.

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$$Q(f) = \tilde{O}(n^{1-2^{-d}})_{\text{[new]}}$$

$$Q(f) = \tilde{O}(n^{7/8})_{\text{[new]}}$$

$$Q(f) = \tilde{O}(n^{3/4})_{\text{[Childs, Kothari, Kimmel 2012]}}$$

$$Q(\text{OR}_n) = O(\sqrt{n})_{\text{[Grover96, BBBV97]}}$$

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- Nearly matches known lower bound: for every d , there is a $f \in LC_d^0$ with $Q(f) = \Omega(n^{1-2^{-O(d)}})$. [Childs, Kothari, Kimmel 2012].

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- Algorithmic application: the circuit class LC_d^0 can be agnostically PAC learned in time $2^{\tilde{O}(n^{1-2^{-d}})}$. [Kalai, Klivans, Mansour, Servedio 2005]

Implications for Circuit Lower Bounds

- An important frontier problem in circuit complexity is to show that IP2 cannot be computed by $AC^0 \circ \oplus$ circuits of polynomial size.
- Best known lower bound: IP2 cannot be computed by depth- d $AC^0 \circ \oplus$ circuits of size $n^{1+4^{-d}}$ [CR96, CGJWX16].
- Our upper bound of $Q(f) = \tilde{O}\left(n^{1-2^{-d}}\right)$ for LC_d^0 circuits f implies that IP2 cannot be computed by depth d $AC^0 \circ \oplus$ circuits of size $O(n^{1+2^{-d}})$, even on a $\frac{1}{2}+1/\text{poly}(n)$ fraction of inputs. [cf. Tal17].

Open Questions

Is our Upper Bound on $Q(\text{LC}_d^0)$ Tight Up To Logs?

- Recall: we show $Q(\text{LC}_d^0) = \tilde{O}(n^{1-2^{-d}})$.
- Nearly matches a known lower bound: for every d , there is an LC_d^0 circuit f with $Q(f) = \Omega(n^{1-2^{-O(d)}})$. [Childs, Kothari, Kimmel 2012].

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- To accomplish this, it would be enough to achieve the following:

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- To accomplish this, it would be enough to achieve the following:

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- Alternatively, show that every (quadratic? polynomial?) size DNF has sublinear quantum query complexity!

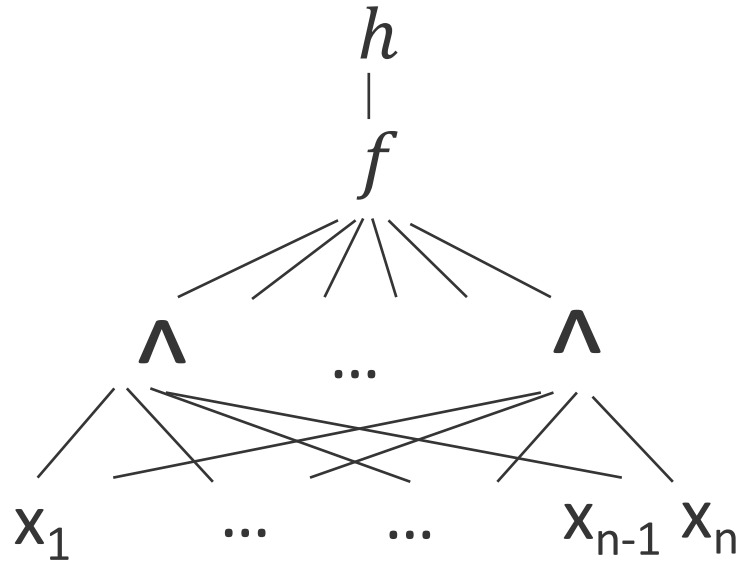
The Lower Bound Argument of [CKK 2012]

- Fact [Beame and Machmouchi 2012, Sherstov 2015]: There is a quadratic size AC^0 circuit C of depth **three**, called SURJECTIVITY, with $Q(C) = \Omega(n)$.
- [CKK 2012] show how to turn C into a **linear-size** circuit C' of depth 3 such that $Q(C') = \Omega(n^{3/4})$.
 - $C' := C_{n^{1/2}} \circ \text{AND}_{n^{1/2}}$.
 - Recall that $Q(C_{n^{1/2}} \circ \text{AND}_{n^{1/2}}) = \Theta\left(Q(C_{n^{1/2}}) \cdot Q(\text{AND}_{n^{1/2}})\right) = \Omega(n^{3/4})$.

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- Then $C'' := C_{n^{1/2}} \circ C'_{n^{1/2}}$ is a linear size circuit of depth 5 such that $Q(C'') = \Omega(n^{7/8})$.
- Then $C''' := C_{n^{1/2}} \circ C''_{n^{1/2}}$ is a linear size circuit of depth 7 such that $Q(C''') = \Omega(n^{15/16})$. And so on.

Generalizing Our Composition Theorem?



Main Theorem: $Q(h) \leq O(\sqrt{Q(f)} \cdot n \cdot \text{polylog}(n)).$

- NOTE: if the roles of AND and f are reversed, the composition theorem is **FALSE**.
- But other generalizations of the composition theorem may be possible.

More Classical Implications of Quantum Query Upper Bounds?

- A sublinear upper bound on the quantum query complexity or approximate degree of a circuit class establishes very strong limitations on that class.
 - Immediately implies the class cannot compute IP_2 , even on average (e.g., [Tal17]).
 - Immediately implies subexponential time agnostic PAC learning algorithms for the class.
 - **Open Question 2: Identify other implications, e.g., Faster SAT algorithms for the class?**

Non-Trivial Query or Communication Upper Bounds for Stronger Circuit Classes?

- Can we show sublinear quantum query or approximate degree upper bounds for circuit classes beyond LC_d^0 ?
 - Maybe this is a path towards showing $IP2 \notin AC^0 \circ \oplus$.
- Alternatively, sublinear **threshold degree** (or subexponential **sign-rank**) upper bounds for a circuit class imply the class cannot compute IP2 in the worst case, and yield PAC learning algorithms for the class.
- Can we show such bounds for interesting circuit classes beyond LC_d^0 ?
 - E.g., showing LT_2 circuits have sub-exponential sign-rank would mean such circuits cannot compute IP2.